

The OT error-driven ranking model of the acquisition of phonotactics

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Phonotactics

- Knowledge of the **phonotactics** of a language is knowledge of the distinction between licit and illicit sounds and sound combinations
- For instance, English speakers know that *blik* could be a licit English word while *bnik* could not, although both are unattested
- Knowledge of **alternations** is knowledge of how to repair illicit sound combinations

The acquisition of phonotactics

Phonotactics is acquired:

- **Early:** nine-month-old infants already react differently to licit and illicit sound combinations (Jusczyk *et al.* 1993)
- **Before morphology:** namely at a stage where the child has still no access to phonological alternations (Hayes 2004)
- **Gradually:** as the target adult phonotactics is approached through a stepwise progression of intermediate stages (McLeod *et al.* 2001)

2:3	2:5	2:6	2:8	2:8	2:10	2:11	3:1
tɪk	lɪk	dk	flɪk	kɪk	kəɪɪ:k	klɪk	klɪk
tɪk	lɪk	dɪk	θlɪk	kɪk	kɪk	klɪk	klɪk
	flɪkθ	klɪkθ	θlɪk	kɪk	kəɪɪk	klɪks	
	klɪkθ				kəɪɪk	kɪk	

Want a model of the child acquisition of phonotactics that makes sense of these properties

Optimality Theory (OT)

OT is built on three core ideas (Prince & Smolensky 2004, Kager 1999):

- Relevant phonological properties are extracted by a set of universal, innate **constraints**, that measure how structures deviate from the ideal
 - ▶ **Markedness constraints** measure how phonological structures violate wellformedness conditions:
 - *DORSAL is violated by dorsal consonants (i.e. by [kl] and [k])
 - ▶ **Faithfulness constraints** measure how phonological realizations differ from intended targets:
 - MAX is violated by deletion (i.e. by [k] as the realization of target /kl/)
- Two or more constraints can **conflict**
 - ▶ MAX prefers [kl] over [t] as the production of the target /kl/
 - ▶ *DORSAL prefers [t] over [kl] instead
- Grammars differ in how they **rank** constraints and conflicts are resolved by a grammar in favor of the constraint it ranks at the top

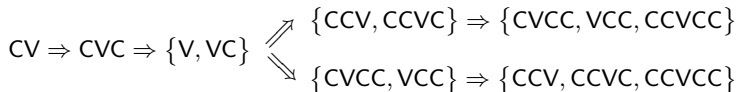
The EDRA model of the acquisition of phonotactics

- **Initialization:** Markedness constraints start at the top, faithfulness constraints at the bottom, yielding the smallest language
- **Loop:** maintains a numerical representation of the current ranking by assigning each constraint a **ranking value** reflecting its relative rank:
 - Step 1 receives a piece of data from the incoming stream of data
 - ▶ receives the word *clock*, that shows that /kl/ is a licit cluster
 - Step 2 checks whether its current ranking accounts for this datum
 - ▶ checks whether it accounts for the mapping /kl/→[kl]
 - ▶ suppose it instead predicts [kl] to be unavailable because reduced to [t]
 - Step 3 if it makes a mistake, it updates the current ranking
 - ▶ slightly demotes *DORSAL that **prefers the loser** mapping /kl/→[t]
 - ▶ slightly promotes MAX, that **prefers the winner** mapping /kl/→[kl]
- **Termination:** loops until constraints (hopefully) intersperse in a ranking consistent with the target phonotactics

Examples: Tesar & Smolensky's (1998) EDCD, Boersma's (1998) GLA.

An example: modeling the acquisition of syllable types

- Attested acquisition paths for Dutch learners (Levelt *et al.* 2001):



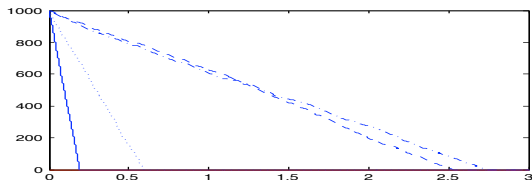
- OT constraints for syllable types (Prince & Smolensky 2004):

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- Re-ranking rule (a mixture of EDCD and GLA):

- ▶ Demote by 1 loser-preferring constraints
- ▶ do nothing to winner-preferring constraint

- Ranking dynamics matches acquisition order (Boersma&Levelt 2001):



Strengths of the OT error-driven model

EDRAs have been endorsed by the OT acquisition literature because

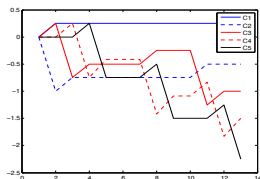
- **Gradualness:** intermediate rankings entertained by EDRAs on the way to the final grammar correspond to intermediate child acquisition stages, thus modeling gradualness
- **Morphology-free:** being trained on faithful mappings ($/kl/ \rightarrow [kl]$), EDRAs only look at surface phonology and don't require alternations, that become available only later on, when morphology kicks in
- **Memory-free:** EDRAs don't keep track of previously seen data, and thus don't impose unrealistic memory requirements (contrary to **batch** models, that are allowed to glimpse at the entire set of data at once).

Main goal of my current research:

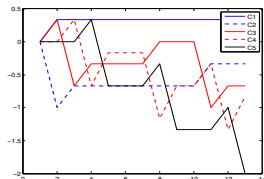
show that EDRAs provide a proper model of the acquisition of phonotactics, both from a computational and a modeling perspective

A combined computational/modeling approach

- Establishing the adequacy of EDRAs looks like an empirical **modeling** issue: you test them on lots of data, as Levelt's syllable types data
- Empirical it is indeed. Yet, here is a way to appreciate the importance of a solid **computational** understanding of the model:



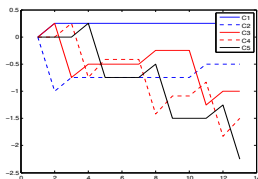
promote by $p - \epsilon$



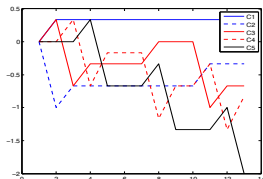
promote by p

A combined computational/modeling approach

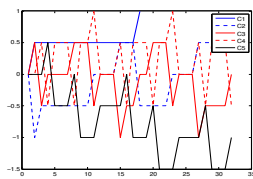
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promote by $p - \epsilon$



promote by p



promote by $p + \epsilon$

- We need an integrated computational/modeling approach:
 - ▶ individuate specific computational desiderata
 - ▶ develop implementations of EDRAs that provably satisfy them
 - ▶ test the modeling adequacy of these EDRAs on child acquisition data
- This project contributes to a new field of linguistic research, called **Cognitive Computational Phonology**

Core issues of the theory of OT error-driven learning

- Issue #1 does the model eventually stop making mistakes and settle on a final grammar? (**convergence**)
- Issue #2 does the final grammar entertained by the model indeed capture the target phonotactics? (**correctness**)
- Issue #3 do the learning sequences predicted by the model match attested child acquisition paths? (**modeling adequacy**)
- Issue #4 how does the model behave in the presence of noise and how can it make sense of child variation? (**robustness/variation**)
- Issue #5 how can the choice of the OT framework be justified from a learning theoretic perspective? (**framework selection**)

The issue of convergence

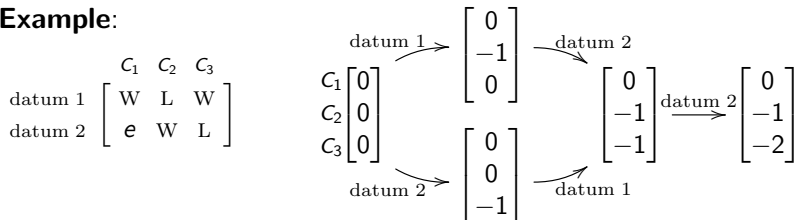
- The most basic computational desiderata on EDRA is:
 - ▶ **Convergence**: is it the case that the EDRA can only make a finite number of mistakes (when trained on consistent data)?
 - ▶ **Efficiency**: is it the case that the number of mistakes grows slowly with the complexity of the underlying typology (= number of constraints)?
- For simplicity, assume from now on that each piece of data has a unique loser-preferrer (this is not a restrictive assumption)
- The prototypical EDRA re-ranking rule works as follows:
 - ▶ **Demotion component**: decrease the ranking value of the unique loser-preferrer by a certain amount, say 1 for concreteness
 - ▶ **Promotion component**: increase the ranking value of each one of the winner-preferrers by a certain promotion amount, call it $p \geq 0$
- The crucial issue is how to choose the promotion amount p :
 - ▶ which choices are needed for efficient convergence?
 - ▶ do they yield good modeling predictions?

The choice $p = 0$: Tesar & Smolensky's (1998) analysis

■ Idea: to be on the safe side, we only perform demotion

- ▶ demote the loser-preferrer by 1
- ▶ do not promote any constraints

■ Example:



■ Sketch of T&S's analysis:

- ▶ C_1 stays at 0; C_2 never makes it below -1 ; C_3 never below -2
- ▶ At each update, one constraint is demoted
- ▶ As constraints cannot drop too much, then you can't update too much
- ▶ I.e., the number of updates is at most $0 + 1 + 2 = \frac{1}{2}n(n-1)$, $n = 3$

The choice $p = 0$: Tesar & Smolensky's analysis (cont'd)

- Each set of OT-compatible data has the shape represented here, by reordering the data and possibly relabeling the constraints
- C_1 stays at 0, C_2 never makes it below -1 , C_3 never below -2 , ..., C_n never below $n-1$
- At each update, one constraint is demoted
- As constraint cannot drop too much, then you can't update too much
- I.e., the number T of updates is at most

$$\begin{aligned}
 T &\leq \begin{matrix} \# \text{ of times} \\ C_1 \text{ is demoted} \end{matrix} + \begin{matrix} \# \text{ of times} \\ C_2 \text{ is demoted} \end{matrix} + \dots + \begin{matrix} \# \text{ of times} \\ C_n \text{ is demoted} \end{matrix} \\
 &\leq 0 + 1 + \dots + (n-1) = \frac{1}{2}n(n-1)
 \end{aligned}$$

C_1	C_2	C_n
w					
w					
e	w				
e	w				
					
					
e	e	-	e	w	
e	e	-	e	w	

Theorem

T&S's demotion-only EDRA performs at most $\frac{1}{2}n(n-1)$ updates (where n is the number of constraints)

But we want constraint promotion!

- The model always assumes the winner mapping is the faithful one (/kI/ → [kI]), which is the safe option if you are blind to alternations
- This means the faithfulness constraints are never loser-preferrers and thus are never re-ranked by T&S's demotion-only EDRA
- This cannot be right:
 - ▶ the demotion-only EDRA fails on target languages that require a different relative ranking of the faithfulness constraints
 - ▶ the demotion-only EDRA fails to model learning paths where the repair changes over time, requiring re-ranking of the faithfulness constraints
- Thus constraint promotion is needed from a **modeling perspective**
- Unfortunately, constraint promotion is not easy to get from a **computational perspective**: Tesar & Smolensky warn us against it!

The choice $p = 1$: Boersma's (1998) attempt

- **Idea:** if we can't choose which to promote, let's just promote all
 - ▶ Demote the loser-preferrer by 1
 - ▶ promote each winner-preferrer by 1
- But Pater (2008) reports simulation results with the following data sampled uniformly where the EDRA does not converge

	C_1	C_2	C_3	C_4	C_5
1	W	L	W		
2		W	L	W	
3			W	L	W
4				W	L

- Actually, the EDRA never converges on these data, no matter how the data are sampled
- How could we reconcile these two opposite perspectives?
 - ▶ the **modeling perspective**, that wants constraint promotion
 - ▶ the **computational perspective**, that wants demotion-only

The choice $p < 1/w$: calibrated constraint promotion

- **Idea:** calibrate promotion, so as not to disrupt convergent demotion
 - ▶ Demote the loser-preferrer by 1
 - ▶ promote each of the w winner-preferrers by less than $\frac{1}{w}$, say by $\frac{1}{w+1}$
- **Step 1 of the analysis:**
 - ▶ C_1 stays at 0, C_2 never makes it below -1 , C_3 never below -2 , ...
 - ▶ hence the sum of current ranking values is always at least $-\frac{1}{2}n(n-1)$
- **Step 2 of the analysis:**
 - ▶ Suppose the current datum has $w = 3$ winner-preferrers
 - ▶ thus the sum of the ranking values is decreased by 1 (demotion)
 - ▶ and increased three times by $\frac{1}{w+1} = \frac{1}{4}$ (promotion)
 - ▶ overall it is therefore decreased by $\frac{1}{4} = \frac{1}{w+1} \geq \frac{1}{n}$
- **Step 3 of the analysis:** updates cannot go on for ever, as the sum of the ranking values shrinks with every update but cannot get too small

Theorem

The well calibrated promotion/demotion EDRA performs at most $\frac{1}{2}n^2(n-1)$ updates (where n is the number of constraints)

The choice $p = 1/w$: the breakpoint

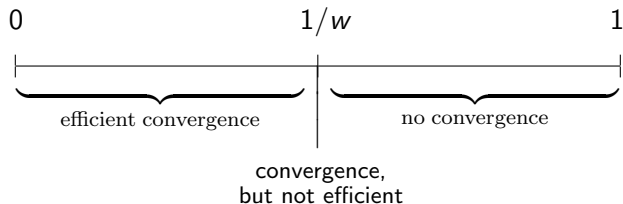
- **Idea:** what if we promote by the smallest non calibrated amount?
 - ▶ Demote the loser-preferrer by 1
 - ▶ promote each of the w winner-preferrers by exactly $\frac{1}{w}$
- The previous analysis of convergence does not extend to this case:
 - ▶ Suppose the current datum has $w = 3$ winner-preferrers
 - ▶ thus the sum of the ranking values is decreased by 1 (demotion)
 - ▶ and increased three times by $\frac{1}{w} = \frac{1}{3}$ (promotion)
 - ▶ overall, the sum of the ranking values remains constant
- Alternative analysis I: use convergence of the Perceptron algorithm
- Alternative analysis II:
 - ▶ suppose the data are **OT-compatible**
 - ▶ this entails that the update vectors are **conically independent**
 - ▶ this entails in turn that the EDRA **cannot loop**
 - ▶ and this entails in turn that the EDRA **converges**

Theorem

The EDRA with $p = \frac{1}{w}$ converges. Yet, efficiency is lost, as the worst-case number of mistakes grows exponentially in the number n of constraints.

Conclusion on convergence

- From a **modeling perspective**, we need EDRA that perform constraint promotion too
- From a **computational perspective**, efficiently convergent constraint promotion is possible through proper calibration



- \implies OT grammars must be represented by the learner as **numerical** ranking values not as **combinatorial** rankings, as calibration of constraint promotion requires a numerical representation of rankings

The issue of correctness

- Phonotactics is the knowledge of licit vs illicit sound combinations
- A grammar (i.e. an OT ranking) is called
 - ▶ **consistent**, provided it rules in every licit form
 - ▶ **restrictive**, provided it also rules out every illicit form
 - ▶ **correct**, provided it is both consistent and restrictive
- Now let's look at EDRA's:
 - ▶ if it converges, then its final grammar is consistent
 - ▶ yet, it could be non-restrictive (e.g. all faithfulness constraints at top)
 - ▶ a convergent EDRA is **correct** provided that the final grammar entertained at converge is also restrictive, and thus correct
- Correctness is a pressing issue for error-driven learning:
 - ▶ we only have control on the initial ranking and the re-ranking rule
 - ▶ thus, the acquisition path is governed by the stream of data
 - ▶ so that the model behaves as a leaf in the wind of data
 - ▶ and there seem to be no reasons to expect correctness
- I want to show that this pessimism is not warranted

My research strategy on EDRA's correctness

■ Step 1 of the analysis:

- ▶ a language is called \mathcal{F} -irrelevant if the relative ranking of the faithfulness constraints does not matter for that language
- ▶ EDRA's are always correct on \mathcal{F} -irrelevant languages, with no restrictions on the constraints

■ Step 2 of the analysis:

- ▶ The problem of the acquisition of phonotactics is not solvable in its full generality, without restrictions on constraints
- ▶ Thus, correctness on \mathcal{F} -relevant languages cannot be achieved without restrictions on constraints (by EDRA's or by any other algorithm)

■ Step 3 of the analysis: investigate the algorithmic implications of generalizations on phonologically plausible constraints for error-driven correctness on \mathcal{F} -relevant languages:

- ▶ assemble a very large constraint set from the OT literature
- ▶ extract \mathcal{F} -relevant languages from the corresponding typology
- ▶ study behavior of EDRA's on those

\mathcal{F} -irrelevant languages

- **Informally:** a language is \mathcal{F} -irrelevant provided the relative ranking of the faithfulness constraints does not matter for that language
- **Formally:**
 - ▶ a **partial ranking** is a partial order of the constraint set: there might be two constraints that are not ranked relative to each other
 - ▶ a partial ranking **generates** a language provided each one of its total refinements generates that language (in the usual OT sense)
 - ▶ a language is **\mathcal{F} -irrelevant** provided it is generated by a partial ranking that doesn't rank any two faithfulness constraints relative to each other
- **Remarks:**
 - ▶ there are partial rankings that don't generate any language (as they admit total refinements that generate different languages)
 - ▶ certain ranking conditions are crucial for consistency (if you drop them, then you allow for total refinements that generate smaller languages)
 - ▶ certain ranking conditions are crucial for restrictivity (if you drop them, then you allow for total refinements that generate larger languages)

Examples

- The OT typology:

$$\begin{array}{l}
 \text{forms} = \left\{ \begin{array}{cccc} \text{ap} & \text{ab} & \text{as} & \text{az} \\ \text{apsa} & \text{absa} & \text{apza} & \text{abza} \end{array} \right\} \\
 \text{constraints} = \left\{ \begin{array}{ll} F_1 = \text{IDENT}[\text{FRIC-VOI}] & F_2 = \text{IDENT}[\text{STP-VOI}] \\ M_1 = *[\text{+FRIC-VOI}] & M_2 = *[\text{+STP-VOI}] \\ M = \text{AGREE}[\text{STP-VOI, FRIC-VOI}] & \end{array} \right\}
 \end{array}$$

- A partial ranking and some of its total refinements:

$$\begin{array}{ccc}
 \begin{array}{cc} M & \\ | & \\ F_1 & F_2 \\ | & | \\ M_1 & M_2 \end{array} & \Rightarrow & \begin{array}{ccc} F_2 & M & M \\ | & | & | \\ M & F_2 & F_2 \\ | & | & | \\ M_2 & F_1 & M_2 \\ | & | & | \\ F_1 & M_2 & F_1 \\ | & | & | \\ M_1 & M_1 & M_1 \end{array}
 \end{array}$$

- This partial ranking thus generates L , which is therefore \mathcal{F} -irrelevant:

$$L = \left\{ \begin{array}{cccc} \text{pa,} & \text{ba,} & \text{sa,} & \text{za,} \\ \text{apsa,} & & & \text{abza,} \end{array} \right\}$$

Examples (cont'd)

- Again, the same OT typology:

$$\begin{array}{l}
 \text{forms} = \left\{ \begin{array}{cccc} \text{ap} & \text{ab} & \text{as} & \text{az} \\ \text{apsa} & \text{absa} & \text{apza} & \text{abza} \end{array} \right\} \\
 \text{constraints} = \left\{ \begin{array}{ll} F_1 = \text{IDENT}[\text{FRIC-VOI}] & F_2 = \text{IDENT}[\text{STP-VOI}] \\ M_1 = *[\text{+FRIC-VOI}] & M_2 = *[\text{+STP-VOI}] \\ M = \text{AGREE}[\text{STP-VOI, FRIC-VOI}] & \end{array} \right\}
 \end{array}$$

- The following partial ranking admits all preceding total refinements:

$$\begin{array}{ccc}
 M & & F_1 \\
 & & | \\
 & & F_2 \\
 & & | \\
 F_1 & F_2 & \Rightarrow & M \\
 | & | & & | \\
 M_1 & M_2 & & M_2 \\
 & & & | \\
 & & & M_1
 \end{array}$$

- But it also admits the total refinement above that generates the entire set of forms, so that this partial ranking generates nothing
- This shows that some ranking conditions are needed for **restrictiveness**

Examples (cont'd)

- Again, the same OT typology:

$$\begin{array}{l}
 \text{forms} = \left\{ \begin{array}{cccc} \text{ap} & \text{ab} & \text{as} & \text{az} \\ \text{apsa} & \text{absa} & \text{apza} & \text{abza} \end{array} \right\} \\
 \text{constraints} = \left\{ \begin{array}{ll} F_1 = \text{IDENT}[\text{FRIC-VOI}] & F_2 = \text{IDENT}[\text{STP-VOI}] \\ M_1 = *[\text{+FRIC-VOI}] & M_2 = *[\text{+STP-VOI}] \\ M = \text{AGREE}[\text{STP-VOI}, \text{FRIC-VOI}] & \end{array} \right\}
 \end{array}$$

- The following partial ranking admits all preceding total refinements:

$$\begin{array}{ccc}
 M & & M \\
 | & & | \\
 F_2 & F_1 & M_1 \\
 & | & | \\
 M_1 & M_2 & F_1 \\
 & & | \\
 & & F_2 \\
 & & | \\
 & & M_2
 \end{array}
 \Rightarrow$$

- But it also admits the total refinement above that is not consistent with the form [ab] and thus does not generate the language L
- This shows that some ranking conditions are needed for **consistency**

EDRAs get right the ranking of M above F

- Suppose that the target language
 - ▶ is \mathcal{F} -irrelevant, namely is generated by a partial ranking that does not rank faithfulness constraints relative to each other
 - ▶ and this partial ranking ranks some M above some F

- This entails that:

- ▶ only markedness constraints can be ranked above M , not any faithfulness constraint
- ▶ the top-ranked one(s) among the markedness constraints ranked above M is not violated by any form in the language

M'
|
 M''
|
⋮
 M
|
 F

- We can thus reason as follows:

- ▶ there can be at most $n - 2$ markedness constraints ranked above M
- ▶ T&S thus guarantee that M cannot drop by more than $n - 3$
- ▶ suppose the EDRA only demotes or else does not promote “too much”
- ▶ so that it won’t promote F that high from its initial low position
- ▶ the EDRA will thus converge to a final ranking that enforces $M > F$

- This reasoning fails if M needs to be ranked underneath some other F' , as could be the case if the language is not \mathcal{F} -irrelevant

EDRAs get right the ranking of M above M'

- Suppose that the target language
 - ▶ is generated by a partial ranking that ranks M above M'
 - ▶ and this ranking happens to be crucial in particular for **restrictiveness**

- This means that:

- ▶ there is a form x not in the language
- ▶ because reduced to some form y
- ▶ and M is the “unprotected” L
- ▶ while M' is a w that could protect M'
- ▶ that thus needs to be ranked below M

$$\begin{array}{l} \dots M \dots M' \dots \\ /x/ \rightarrow [x]: [\dots L \dots W \dots] \\ /x/ \rightarrow [y]: \end{array}$$

- Now we can reason as follows:

- ▶ the form y must be in the language
- ▶ suppose “candidacy” is symmetric
- ▶ so that x is a candidate of y
- ▶ now M' is an L that needs to be protected with the w of M
- ▶ so that M needs to be ranked above M' also for **consistency**
- ▶ as convergent EDRAs have no problems with consistency
- ▶ the final grammar entertained by convergent EDRAs enforces $M > M'$

Step 1 of correctness: \mathcal{F} -irrelevant languages

Summarizing:

F cannot matter, if the language
|
 F' is \mathcal{F} -irrelevant

F can only matter for con-
|
 M consistency, so that convergent
EDRAs get it right

M EDRAs get it right, if the lan-
|
 F guage is \mathcal{F} -irrelevant and pro-
motion is null or calibrated

M it cannot matter only for re-
|
 M' strictiveness provided “candi-
dacy” is symmetric, so that
EDRAs get it right

Thus, we conclude that:

Theorem (Step 1 of the analysis of correctness)

If “candidacy” is symmetric:

- demotion-only EDRAs are correct on any \mathcal{F} -irrelevant language
- the result extends to EDRAs that do not promote too much
- but not to EDRAs that promote too much (e.g. Boersma’s GLA)

Step 2 of correctness: \mathcal{F} -relevant languages

- Convergence can be achieved in full generality, namely with no restrictions whatsoever on the underlying OT constraint set
- Correctness on \mathcal{F} -irrelevant languages can be achieved in full generality too (only need “candidacy” to be symmetric)
- But correctness on \mathcal{F} -relevant languages provably cannot be achieved in full generality, no matter the algorithm we pick

Theorem (Step 2 of the analysis of correctness)

The problem of the acquisition of phonotactics in OT cannot be solved efficiently in its general formulation (it is NP-complete), namely without restrictive assumptions on the underlying OT constraint set

- Is it the case that phonologically plausible restrictions on constraints ensure correctness of EDRA on \mathcal{F} -relevant languages?
- A positive answer would provide formidable support for EDRA as a proper model of the child’s acquisition of phonotactics

A couple of cases of segmental OT phonotactics

■ Obstruent voicing (Lombardi 1999, Prince & Tesar 2004):

▶ $forms = \left\{ \begin{array}{cccc} ap & ab & as & az \\ apsa & absa & apza & abza \end{array} \right\}$

▶ $constraints = \left\{ \begin{array}{ll} F_1 = IDENT[FRIC-VOI] & F_2 = IDENT[STP-VOI] \\ M_1 = * [+FRIC-VOI] & M_2 = * [+STP-VOI] \\ M = AGREE[STP-VOI, FRIC-VOI] \end{array} \right\}$

■ Laryngeal inventory (Hayes 2004):

▶ $forms = \left\{ t \quad d \quad t^h \quad d^h \right\}$

▶ $constraints = \left\{ \begin{array}{ll} F_1 = IDENT[VOICE] & F_2 = IDENT[ASP] \\ M_1 = * [+VOICE] & M_2 = * [+ASP] \\ M = * [+VOICE, +ASP] \end{array} \right\}$

Extracting a toy framework for OT segmental phonotactics

- There are two partial binary phonological features

$$\varphi_1, \varphi_2 : \text{forms} \mapsto \{0, 1, \#\}$$

- Segments are pairs of feature values:

$$\mathbf{x} = \langle x_1, x_2 \rangle, x_i \in \{0, 1, \#\}$$

- Candidates obtained by changing feature values of in all possible ways
- The constraint set can contain:

- faithfulness constraint* F_i corresponding to feature φ_i

$$F_i(\mathbf{x}, \mathbf{y}) = 1 \iff x_i \neq y_i$$

- (simple) markedness constraint* M_i corresponding to feature φ_i

$$M_i(\mathbf{y}) = 1 \iff y_i = 1 = \text{the marked value}$$

- binary markedness constraint* M^μ with *markedness pattern* μ :

$$M^\mu(\mathbf{y}) = 1 \iff \langle y_i, y_j \rangle \in \mu$$

- There are sixteen possible markedness patterns μ . Some examples:

- $\mu = \{\langle 1, 1 \rangle\}$

- $\mu = \{\langle 0, 1 \rangle, \langle 1, 0 \rangle\}$

- $\mu = \{\langle 0, 0 \rangle, \langle 1, 1 \rangle\}$

*[+VOICE, +ASP]

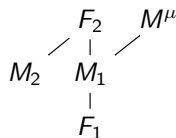
AGREE[STP-VOI, FRIC-VOI]

OCP-type constraint

Phonologically plausible binary markedness constraints

- A binary markedness constraint is **phonologically plausible** provided:
 - ▶ its markedness pattern does not have cardinality 3
 - ▶ its markedness pattern is not the singleton $\{ \langle 0, 0 \rangle \}$
- For each phonologically plausible markedness patterns and each sets of form (with some mild technical restrictions), construct the corresponding typology \implies a total of roughly 150 languages
- There are only 6 languages that are \mathcal{F} -relevant! here I list those that require F_2 above F_1 (the others are analogous by feature symmetry)

markedness pattern μ	target language L
$\mu = \{ \langle 0, 0 \rangle, \langle 1, 1 \rangle \}$	$L = \left\{ \begin{array}{l} \langle 0, 1 \rangle \\ \langle 1, 0 \rangle \end{array} \right\} \cup \left\{ \begin{array}{l} \langle 0, \# \rangle \langle \#, 0 \rangle \\ \langle \#, 1 \rangle \end{array} \right\}$
$\mu = \{ \langle 0, 1 \rangle, \langle 1, 0 \rangle \}$	$L = \left\{ \begin{array}{l} \langle 0, 0 \rangle \\ \langle 1, 1 \rangle \end{array} \right\} \cup \left\{ \begin{array}{l} \langle 0, \# \rangle \langle \#, 0 \rangle \\ \langle \#, 1 \rangle \end{array} \right\}$
$\mu = \{ \langle 0, 1 \rangle \}$	$L = \left\{ \begin{array}{l} \langle 0, 0 \rangle \\ \langle 1, 1 \rangle \end{array} \right\}$
$\mu = \{ \langle 0, 1 \rangle \}$	$L = \left\{ \begin{array}{l} \langle 0, 0 \rangle \\ \langle 1, 1 \rangle \end{array} \right\}$



- This shows that \mathcal{F} -irrelevant languages are the majority, as expected; Step 1 of the analysis of correctness is thus an interesting result

EDRAs on \mathcal{F} -relevant languages for plausible constraints

- Let's look for instance at the first one of these languages:

	F_1	F_2	M_1	M_2	M
$\langle \#,1 \rangle, \langle \#,0 \rangle$		W		L	
$\langle 1,0 \rangle, \langle 0,0 \rangle$	W		L		W
$\langle 1,0 \rangle, \langle 0,1 \rangle$	W	W	L	W	
$\langle 0,1 \rangle, \langle 0,0 \rangle$		W		L	W
$\langle 0,1 \rangle, \langle 1,0 \rangle$	W	W	W	L	

- Data that have a W corresponding to M will trigger at most one update and can thus be effectively discarded:

	F_1	F_2	M_1	M_2	M
$\langle \#,1 \rangle, \langle \#,0 \rangle$		W		L	
$\langle 1,0 \rangle, \langle 0,1 \rangle$	W	W	L	W	
$\langle 0,1 \rangle, \langle 1,0 \rangle$	W	W	W	L	

- If the first datum is fed at least once, the EDRA will get right the desired ranking F_2 above F_1
- Analogous considerations hold for the other \mathcal{F} -relevant languages

Step 3 of correctness: toy version, and beyond

Theorem (Toy version of step 3 of the analysis of correctness)

Within the toy framework for OT segmental phonotactics considered, EDRA's are correct on \mathcal{F} -relevant languages provided (binary markedness) constraints are phonologically plausible

Now I need to scale up this little case study:

- assemble a very large constraint set from the OT literature:
 - ▶ lots of features
 - ▶ markedness constraints that target more than two features
 - ▶ lots of feature interaction (interaction graph)
- extract \mathcal{F} -relevant languages from the corresponding typology
 - ▶ by developing results on licit immediate rankings in **minimal** generating partial rankings
- study behavior of EDRA's on those languages
 - ▶ by developing proper tools for the analysis of EDRA's